In the last 10 years, CPU clock speeds haven’t increased significantly, but the computing power of CPUs has continued to grow according to Moore’s law. Researchers achieved this by increasing the number of computing cores per chip. This process has made parallel computing relevant to a much broader community.

Writing computer programs that take advantage of multicores is a challenge because of the diversity of architectures and because of the inherent difficulty of writing parallel programs. In today’s multicore world, the means to express parallelism and the effectiveness in achieving parallel performance are increasingly important.

Traditionally, there have been two approaches to parallel programming: threads that communicate via shared memory and processes that...
communicate via message passing. Modern programming languages introduce different approaches, some of which are discussed in this special issue of CiSE.

About the Articles

In “Concurrency and Message Passing in Erlang,” Martin Kalin and David Miller describe the Erlang language, which was first released to the public by Rich Hickey in 2007. Erlang—like Lisp, Erlang, and Haskell—is a functional language. In functional languages, the output of any function depends only on the input, and there’s no global state. This approach alone should make functional languages appealing to scientists. The input and output consist of immutable data structures that can safely be accessed concurrently without the need for locking.

If a problem is decomposable, the functions that constitute the intermediate steps to the solution can be executed in parallel, and the underlying virtual machine (VM) can automatically allocate them to different cores. Yet not everything is always so simple and some objects must be mutable. Different languages deal with mutability in different ways. A truly unbiased speed comparison is impossible, and real-life performance is problem-specific. Haskell doesn’t use a VM, but it compiles to native code. Like Erlang, it supports lightweight threads, but it doesn’t use explicit message passing. Instead, it determines data dependencies from the data structures. For example, a Haskell library called Repa defines the notion of a parallel array. Parallel array operations are automatically parallelized into as many chunks as there are CPU cores.

Coutts and Löh also mention ongoing development efforts to let Haskell support grids, GPUs, and CPU single-instruction, multiple-data (SIMD) instructions. The goal is to allow different parallel programming models within a single programming language.

A Parallel Assignment

To better compare Clojure, Erlang, and Haskell, we assigned the authors the same problem: parallelize a naive solver for a 1D Poisson equation:

$$\nabla^2 \phi(x) = \rho(x).$$

Once discretized ($x \rightarrow x_i = x_0 + h_i, \phi(x) \rightarrow \phi_i, \rho(x) \rightarrow \rho_i$) this equation turns into an iterative expression:

$$\forall i \quad \phi_i = \frac{\phi_{i+1} + \phi_{i-1}}{2} - \frac{h^2}{2} \rho_i,$$

where $\rho$ and $\phi$ can be thought of as input and output, respectively. Here, $h$ is the discretization step. The expression must be iterated until convergence of the array $\phi$. Each iteration can be parallelized in the integer variable $i$.

There are three issues of importance here: ease of implementation of the parallel approach; parallel scaling with the number of cores; and overall language speed. We leave the discussion of the first two issues to the authors.

A truly unbiased speed comparison is impossible, and real-life performance is problem-specific. Still, we refer the reader to The Computer Language Benchmarks Game (http://shootout.alioth.debian.org), from which we report benchmarks (see Table 1). The first column of Table 1 shows the average running time of a battery of benchmarks on a single core of a Q6600 Intel CPU normalized to C++ running time. This measures the language speed but not its scaling. The second column shows the results of a specific benchmark, called thread-ring, on four cores. Thread-ring creates 503 threads and passes a token around them in a ring, while performing no computation. This measures the language overhead in dealing with concurrency. In both cases, smaller numbers indicate better performance.
The proliferation of new languages and paradigms can at first appear confusing, but it shows a clear trend toward increasing the expressiveness and readability of languages while decoupling the coding of the algorithm from parallelization and concurrency optimizations.

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